# Class 9: Halloween Mini-Project

Sarah Mirsaidi Madjdabadi, A16890186

# Table of contents

1. Importing candy data	1
2. What is your favorate candy?	2
Exploratory Analysis	3
Overall Candy Rankings	5
4. Taking a look at pricepercent	11
5. Correlation Structure	13
6. Principal Component Analysis (PCA)	15

# 1. Importing candy data

```
candy <- read.csv("candy-data.txt", row.names=1)
head(candy)</pre>
```

	choco	olate	fruity	caramel	peanutyalmondy	nougat	crispedricewafer
100 Grand		1	0	1	0	0	1
3 Musketeers		1	0	0	0	1	0
One dime		0	0	0	0	0	0
One quarter		0	0	0	0	0	0
Air Heads		0	1	0	0	0	0
Almond Joy		1	0	0	1	0	0
	${\tt hard}$	bar j	pluribus	sugarpe	ercent priceper	cent wir	npercent
100 Grand	0	1	0	)	0.732 0	.860	66.97173
3 Musketeers	0	1	0	)	0.604 0	.511 6	67.60294
One dime	0	0	0	)	0.011 0	.116 3	32.26109
One quarter	0	0	C	)	0.011 0	.511 4	46.11650
Air Heads	0	0	C	)	0.906 0	.511 5	52.34146
Almond Joy	0	1	0	)	0.465 0	.767	50.34755

Q1. How many different candy types are in this dataset?

```
nrow(candy)
```

[1] 85

There are 85 different candy types in this data set.

Q2. How many fruity candy types are in the dataset?

### table(candy\$fruity)

0 1

47 38

There are 38 fruity candy types in the dataset.

# 2. What is your favorate candy?

Q3. What is your favorite candy in the dataset and what is it's winpercent value?

```
candy["Twix", ]$winpercent
```

[1] 81.64291

Q4. What is the winpercent value for "Kit Kat"?

```
candy["Kit Kat", ]$winpercent
```

[1] 76.7686

Q5. What is the winpercent value for "Tootsie Roll Snack Bars"?

```
candy["Tootsie Roll Snack Bars", ]$winpercent
```

[1] 49.6535

# **Exploratory Analysis**

We can use the **skimr** package to get a quick overview of a given dataset. This can be useful for the first time you encounter a new dataset.

skimr::skim(candy)

Table 1: Data summary

Name	candy
Number of rows	85
Number of columns	12
Column type frequency:	
numeric	12
Group variables	None

### Variable type: numeric

skim_variable n_	_missingcom	plete_ra	tuenean	sd	p0	p25	p50	p75	p100	hist
chocolate	0	1	0.44	0.50	0.00	0.00	0.00	1.00	1.00	
fruity	0	1	0.45	0.50	0.00	0.00	0.00	1.00	1.00	
caramel	0	1	0.16	0.37	0.00	0.00	0.00	0.00	1.00	
peanutyalmondy	0	1	0.16	0.37	0.00	0.00	0.00	0.00	1.00	
nougat	0	1	0.08	0.28	0.00	0.00	0.00	0.00	1.00	
crispedricewafer	0	1	0.08	0.28	0.00	0.00	0.00	0.00	1.00	
hard	0	1	0.18	0.38	0.00	0.00	0.00	0.00	1.00	
bar	0	1	0.25	0.43	0.00	0.00	0.00	0.00	1.00	
pluribus	0	1	0.52	0.50	0.00	0.00	1.00	1.00	1.00	
sugarpercent	0	1	0.48	0.28	0.01	0.22	0.47	0.73	0.99	
pricepercent	0	1	0.47	0.29	0.01	0.26	0.47	0.65	0.98	
winpercent	0	1	50.32	14.71	22.45	39.14	47.83	59.86	84.18	

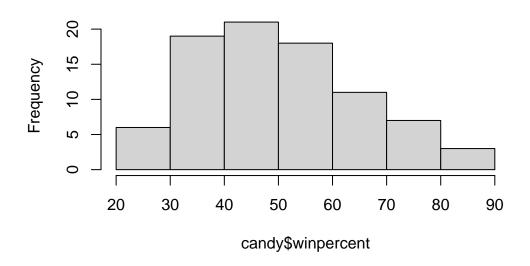
Q6. Is there any variable/column that looks to be on a different scale to the majority of the other columns in the dataset?

Yes, the candy\$winpercent column is on a 0-100 scale while all the others are on a 0-1 scale.

Q7. What do you think a zero and one represent for the candy\$chocolate column? 0 = the candy is not chocolate; 1 = the candy is chocolate.

### hist(candy\$winpercent)

# Histogram of candy\$winpercent



Q9. Is the distribution of winpercent values symmetrical?

No, the distribution is skewed to the left.

Q10. Is the center of the distribution above or below 50%?

### summary(candy\$winpercent)

```
Min. 1st Qu. Median Mean 3rd Qu. Max. 22.45 39.14 47.83 50.32 59.86 84.18
```

The median is below 50%, but the mean (susceptible to outliers) is around 50%.

Q11. On average is chocolate candy higher or lower ranked than fruit candy?

```
choc.inds <- candy$chocolate == 1
choc.candy <- candy[choc.inds, ]
choc.win <- choc.candy$winpercent
mean(choc.win)</pre>
```

#### [1] 60.92153

```
fruity.inds <- candy$fruity == 1
fruity.candy <- candy[fruity.inds, ]
fruity.win <- fruity.candy$winpercent
mean(fruity.win)</pre>
```

#### [1] 44.11974

On average, the chocolate candy is ranked higher than the fruity candy.

Q12. Is this difference statistically significant?

```
ans <- t.test(choc.win, fruity.win)
ans</pre>
```

```
Welch Two Sample t-test
```

```
data: choc.win and fruity.win
t = 6.2582, df = 68.882, p-value = 2.871e-08
alternative hypothesis: true difference in means is not equal to 0
95 percent confidence interval:
    11.44563    22.15795
sample estimates:
mean of x mean of y
    60.92153    44.11974
```

#### ans\$p.value

#### [1] 2.871378e-08

Yes, with a p-value of  $2.8713778 \times 10^{-8}$  (<0.05).

## **Overall Candy Rankings**

Q13. What are the five least liked candy types in this set?

There are two related functions that can help here. One is the classic **sort()** and the other is **order()**.

```
x <- c(5, 10, 1, 4)
sort(x)
```

[1] 1 4 5 10

order(x)

[1] 3 4 1 2

```
inds <- order (candy$winpercent)
head(candy[inds,], 5)</pre>
```

		chocolate	fruity	caran	ו ום	neanutvalm	nondv	nougat	
Nile I Nin		0	11 u1 cy	caran		peanutyain	nonay	nougat	
Nik L Nip	_	-	1		0		0	0	
Boston Baked	Beans	0	0		0		1	0	
Chiclets		0	1		0		0	0	
Super Bubble		0	1		0		0	0	
Jawbusters		0	1		0		0	0	
		crispedrio	ewafer	hard	bar	pluribus	sugar	percent	pricepercent
Nik L Nip			0	0	0	1		0.197	0.976
Boston Baked	Beans		0	0	0	1		0.313	0.511
Chiclets			0	0	0	1		0.046	0.325
Super Bubble			0	0	0	0		0.162	0.116
Jawbusters			0	1	0	1		0.093	0.511
		winpercent	;						
Nik L Nip		22.44534	Ŀ						
Boston Baked	Beans	23.41782	2						
Chiclets		24.52499	)						
Super Bubble		27.30386	3						
Jawbusters		28.12744	Į						

Q14. What are the top 5 all time favorite candy types out of this set?

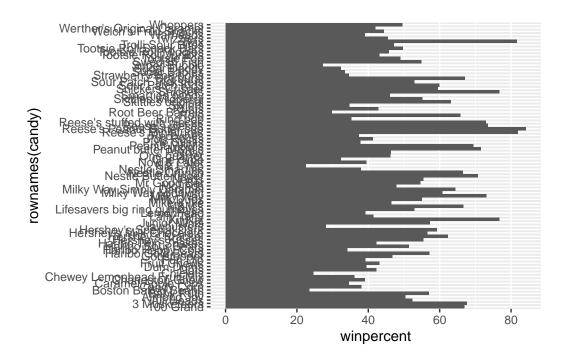
# tail(candy[inds,], 5)

	chocolate	fruity	caramel	peanutyalmondy	nougat
Snickers	1	0	1	1	1
Kit Kat	1	0	0	0	0
Twix	1	0	1	0	0
Reese's Miniatures	1	0	0	1	0

Reese's Peanut Butter c	cup	1	0		0		1	0
		crispedricewa	afer	${\tt hard}$	bar	pluribus	sugarpe	ercent
Snickers			0	0	1	0		0.546
Kit Kat			1	0	1	0		0.313
Twix			1	0	1	0		0.546
Reese's Miniatures			0	0	0	0		0.034
Reese's Peanut Butter c	cup		0	0	0	0		0.720
		pricepercent	winj	percer	ıt			
Snickers		0.651	76	6.6737	78			
Kit Kat		0.511	76	3.7686	60			
Twix		0.906	8:	1.6429	91			
Reese's Miniatures		0.279	8:	1.8662	26			
Reese's Peanut Butter c	cup	0.651	84	1.1802	29			

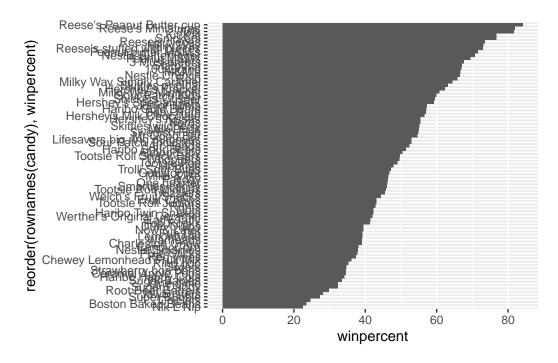
Q15. Make a first barplot of candy ranking based on winpercent values.

```
library(ggplot2)
ggplot(candy) +
  aes(winpercent, rownames(candy)) +
  geom_col()
```



Q16. This is quite ugly, use the reorder() function to get the bars sorted by winpercent?

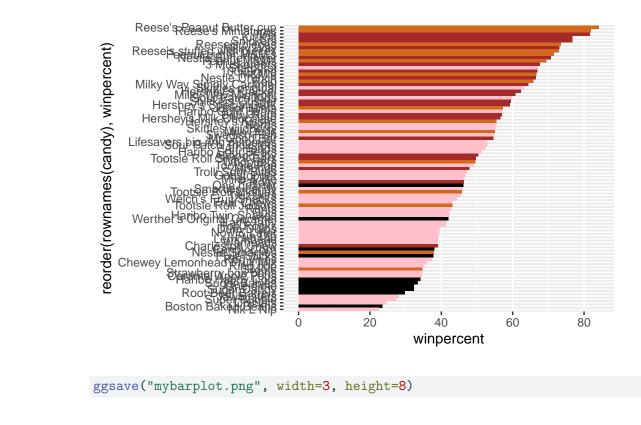
```
ggplot(candy) +
  aes(winpercent, reorder(rownames(candy), winpercent)) +
  geom_col()
```



### Custom Color Vector:

```
my_cols=rep("black", nrow(candy))
my_cols[as.logical(candy$chocolate)] = "chocolate"
my_cols[as.logical(candy$bar)] = "brown"
my_cols[as.logical(candy$fruity)] = "pink"
```

```
ggplot(candy) +
  aes(winpercent, reorder(rownames(candy), winpercent)) +
  geom_col(fill=my_cols)
```



ggsave("mybarplot.png", width=3, height=8)

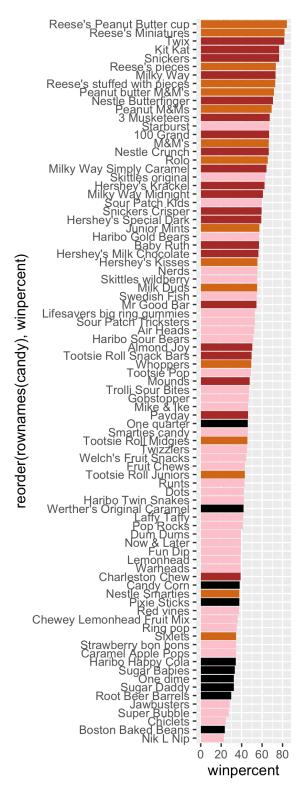


Figure 1: My silly barplot image

Q17. What is the worst ranked chocolate candy?

Sixlets.

Q18. What is the best ranked fruity candy?

Starburst.

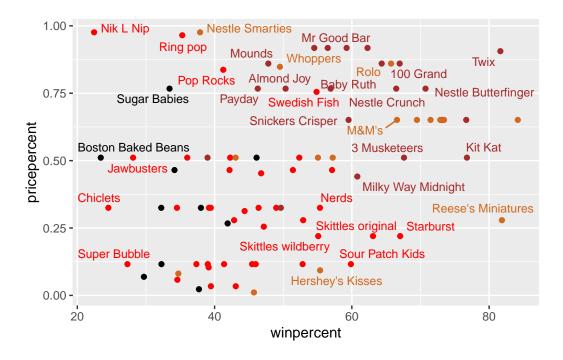
# 4. Taking a look at pricepercent

```
library(ggrepel)

# Pink is too light, let's change to red
my_cols[as.logical(candy$fruity)] = "red"

# How about a plot of price vs win
ggplot(candy) +
   aes(winpercent, pricepercent, label=rownames(candy)) +
   geom_point(col=my_cols) +
   geom_text_repel(col=my_cols, size=3.3, max.overlaps = 8)
```

Warning: ggrepel: 52 unlabeled data points (too many overlaps). Consider increasing max.overlaps



```
ord <- order(candy$pricepercent, decreasing = TRUE)
head( candy[ord,c(11,12)], n=5 )</pre>
```

	pricepercent	winpercent
Nik L Nip	0.976	22.44534
Nestle Smarties	0.976	37.88719
Ring pop	0.965	35.29076
Hershey's Krackel	0.918	62.28448
Hershey's Milk Chocolate	0.918	56.49050

Q19. Which candy type is the highest ranked in terms of winpercent for the least money - i.e. offers the most bang for your buck?

#### Reese's Miniatures.

Q20. What are the top 5 most expensive candy types in the dataset and of these which is the least popular?

Nik L Nip.

#### 5. Correlation Structure

```
cij <- cor(candy)
cij</pre>
```

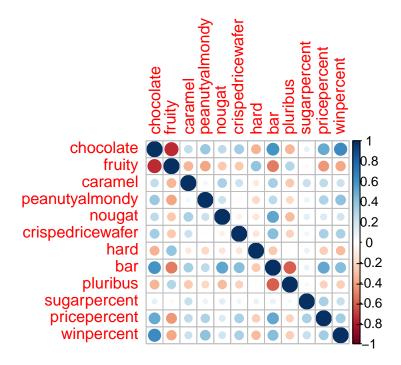
```
chocolate
                                fruity
                                           caramel peanutyalmondy
                                                                       nougat
chocolate
                  1.0000000 -0.74172106
                                        0.24987535
                                                       0.37782357
                                                                   0.25489183
                 -0.7417211 1.00000000 -0.33548538
                                                      -0.39928014 -0.26936712
fruity
caramel
                 0.2498753 -0.33548538
                                        1.00000000
                                                       0.05935614
                                                                   0.32849280
peanutyalmondy
                 0.3778236 -0.39928014
                                        0.05935614
                                                       1.00000000
                                                                   0.21311310
                 0.2548918 -0.26936712
                                        0.32849280
                                                                   1.00000000
nougat
                                                       0.21311310
crispedricewafer
                 0.3412098 -0.26936712
                                        0.21311310
                                                      -0.01764631 -0.08974359
hard
                 -0.3441769 0.39067750 -0.12235513
                                                      -0.20555661 -0.13867505
bar
                 0.5974211 -0.51506558
                                        0.33396002
                                                       0.26041960 0.52297636
pluribus
                 -0.3396752 0.29972522 -0.26958501
                                                      -0.20610932 -0.31033884
sugarpercent
                 0.1041691 -0.03439296
                                        0.22193335
                                                       0.08788927
                                                                   0.12308135
                 0.5046754 -0.43096853
                                        0.25432709
                                                       0.30915323
pricepercent
                                                                   0.15319643
winpercent
                 0.6365167 -0.38093814
                                        0.21341630
                                                       0.40619220 0.19937530
                 crispedricewafer
                                        hard
                                                     bar
                                                            pluribus
chocolate
                      0.34120978 -0.34417691
                                              0.59742114 -0.33967519
                     -0.26936712  0.39067750  -0.51506558
                                                         0.29972522
fruity
                      0.21311310 -0.12235513 0.33396002 -0.26958501
caramel
peanutyalmondy
                     -0.01764631 -0.20555661 0.26041960 -0.20610932
nougat
                     -0.08974359 -0.13867505 0.52297636 -0.31033884
crispedricewafer
                      1.00000000 -0.13867505
                                              0.42375093 -0.22469338
hard
                     -0.13867505
                                  1.00000000 -0.26516504 0.01453172
bar
                      0.42375093 -0.26516504
                                              1.00000000 -0.59340892
pluribus
                     -0.22469338
                                  0.01453172 -0.59340892 1.00000000
sugarpercent
                      0.06994969
                                  0.09180975
                                              0.09998516 0.04552282
pricepercent
                      0.32826539 -0.24436534
                                              0.51840654 -0.22079363
winpercent
                      sugarpercent pricepercent winpercent
chocolate
                  0.10416906
                                0.5046754 0.6365167
                               -0.4309685 -0.3809381
fruity
                  -0.03439296
caramel
                  0.22193335
                                0.2543271 0.2134163
peanutyalmondy
                  0.08788927
                                0.3091532 0.4061922
nougat
                  0.12308135
                                0.1531964 0.1993753
crispedricewafer
                  0.06994969
                                0.3282654
                                           0.3246797
hard
                  0.09180975
                               -0.2443653 -0.3103816
                  0.09998516
bar
                                0.5184065
                                          0.4299293
pluribus
                  0.04552282
                               -0.2207936 -0.2474479
```

sugarpercent	1.00000000	0.3297064	0.2291507
pricepercent	0.32970639	1.0000000	0.3453254
winpercent	0.22915066	0.3453254	1.0000000

# library(corrplot)

corrplot 0.95 loaded

# corrplot(cij)



Q22. Examining this plot what two variables are anti-correlated (i.e. have minus values)?

Fruit and chocolate.

```
round( cij["chocolate", "fruity"], 2)
```

[1] -0.74

Q23. Similarly, what two variables are most positively correlated?

Chocolate and winpercent.

```
round( cij["chocolate", "winpercent"], 2)
```

[1] 0.64

# 6. Principal Component Analysis (PCA)

We need to be sure to scale our input candy data before PCA as we have the winpercent column on a different scale to all others in the dataset.

```
pca <- prcomp(candy, scale=T)
summary(pca)</pre>
```

Importance of components:

```
PC1
                                 PC2
                                        PC3
                                                PC4
                                                       PC5
                                                               PC6
                                                                        PC7
Standard deviation
                       2.0788 1.1378 1.1092 1.07533 0.9518 0.81923 0.81530
Proportion of Variance 0.3601 0.1079 0.1025 0.09636 0.0755 0.05593 0.05539
Cumulative Proportion 0.3601 0.4680 0.5705 0.66688 0.7424 0.79830 0.85369
                           PC8
                                   PC9
                                          PC10
                                                  PC11
                                                          PC12
Standard deviation
                       0.74530 0.67824 0.62349 0.43974 0.39760
Proportion of Variance 0.04629 0.03833 0.03239 0.01611 0.01317
Cumulative Proportion 0.89998 0.93832 0.97071 0.98683 1.00000
```

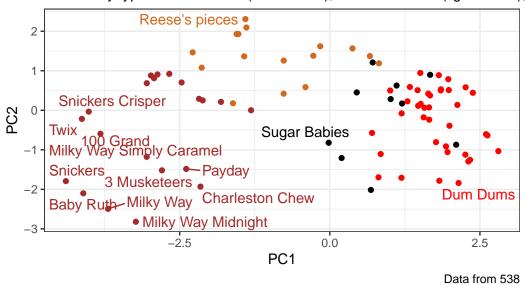
First main result figure is my "PCA plot"

```
ggplot(pca$x) +
  aes(PC1, PC2, label=rownames(pca$x)) +
  geom_point(col=my_cols) +
  geom_text_repel(max.overlaps = 6, col=my_cols) +
  theme_bw() +
  labs(title="Halloween Candy PCA Space",
      subtitle="Colored by type: chocolate bar (dark brown), chocolate other (light brown),
      caption="Data from 538")
```

Warning: ggrepel: 71 unlabeled data points (too many overlaps). Consider increasing max.overlaps

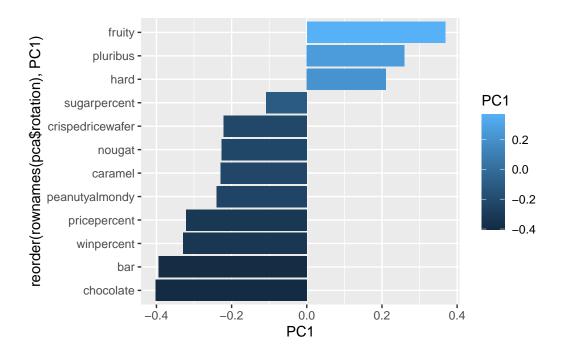
# Halloween Candy PCA Space

Colored by type: chocolate bar (dark brown), chocolate other (light brown),



The second main PCA result is in the pca\$rotation we can plot this to generate a so-called "loadings" plot.

```
ggplot(pca$rotation) +
  aes(PC1, reorder(rownames(pca$rotation), PC1), fill=PC1) +
  geom_col()
```



Q24. What original variables are picked up strongly by PC1 in the positive direction? Do these make sense to you?

Fruity, hard, and pluribus. Yes, these make sense because they are correlated with each other (and not correlated with the other types of candy).